


ROI creation on HLT

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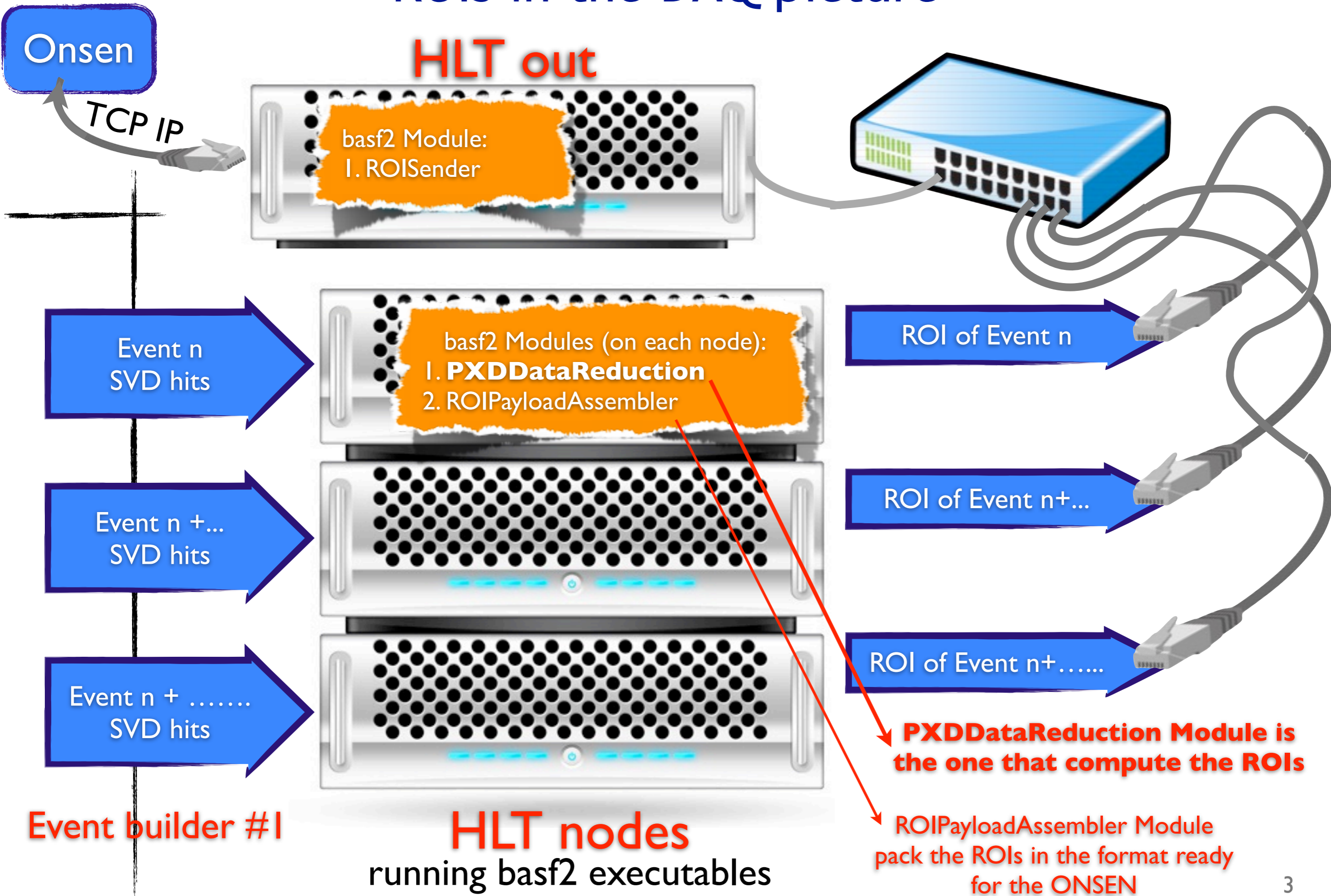
 4th Belle II PXD/SVD Workshop and 14th International
Workshop on DEPFET Detectors and Applications

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Outline

- ➔ ROI basf2 Modules on HLT nodes and HLT output nodes
- ➔ The PXD Data Reduction Module Algorithm
- ➔ Performance of the Module
 - efficiency, data reduction factor, execution time
 - for the Belle2 and the test-beam configurations
- ➔ Conclusions

ROIs in the DAQ picture



Event builder #1

HLT nodes
running basf2 executables

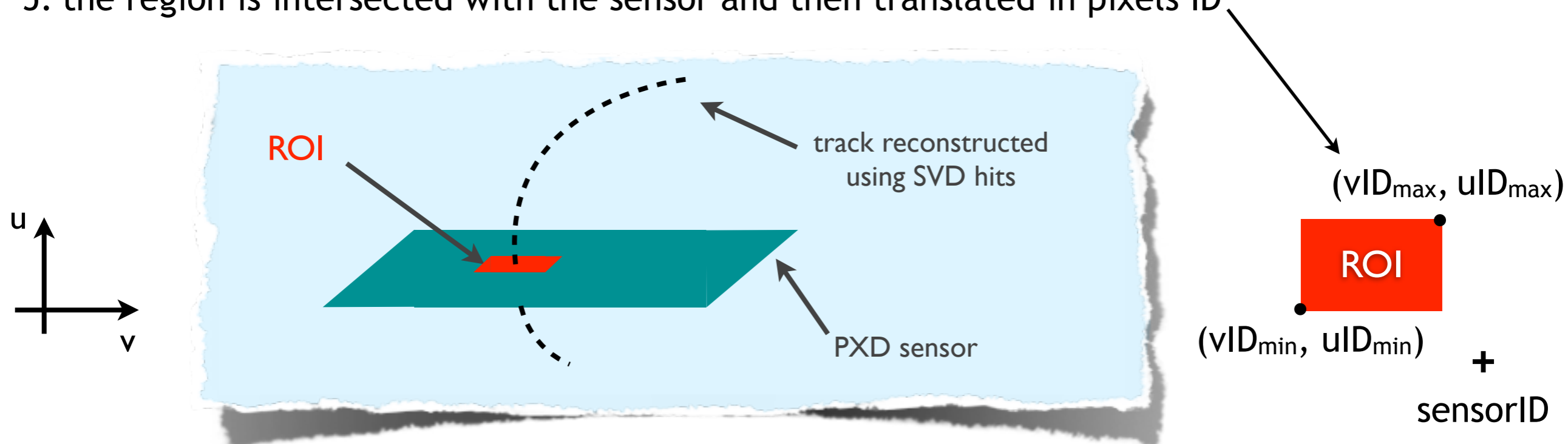
PXDDataReduction Module is the one that compute the ROIs

ROI Payload Assembler Module pack the ROIs in the format ready for the ONSEN

Software-Based Data Reduction Algorithm

[implemented in the `PXDDataReduction`]

1. pattern recognition performed with SVD hits only:
 - ★ GTrackCand list produced by *VXDTF* (or *MCTrackFinder* for testing purposes only)
2. fit the GTrackCand using the standard kalman filter (genfit) and produce a GTrack
 - ★ the fit is done in both directions: first inward, then outward
3. the GTrack is extrapolated on each of the 40 planes containing a PXD sensor
 - ★ obtain an extrapolation point on the plane and the associated statistical errors σ_{stat}
4. a rectangular region is defined given σ_{stat} , a systematic error σ_{syst} and a total number of $\sigma = \text{sqrt}(\sigma_{\text{stat}}^2 + \sigma_{\text{syst}}^2)$ in each direction u, v
5. the region is intersected with the sensor and then translated in pixels ID



ROI Efficiency with MCTrackFinder

- ➔ Definition of efficiency for PXD Data Reduction:

$$\varepsilon = \frac{\text{\# PXDDigits inside a ROI}}{\text{total \# PXDDigits of GFTrackCand}}$$

inefficiencies of the pattern recognition are factorized!!

- ➔ We've simulated 1k events using EvtGen and use the *MCTrackFinder* as pattern recognition:

- ~10 tracks/event
- ~4.2 PXDDigits/track

track candidates are built with the true SVT hits



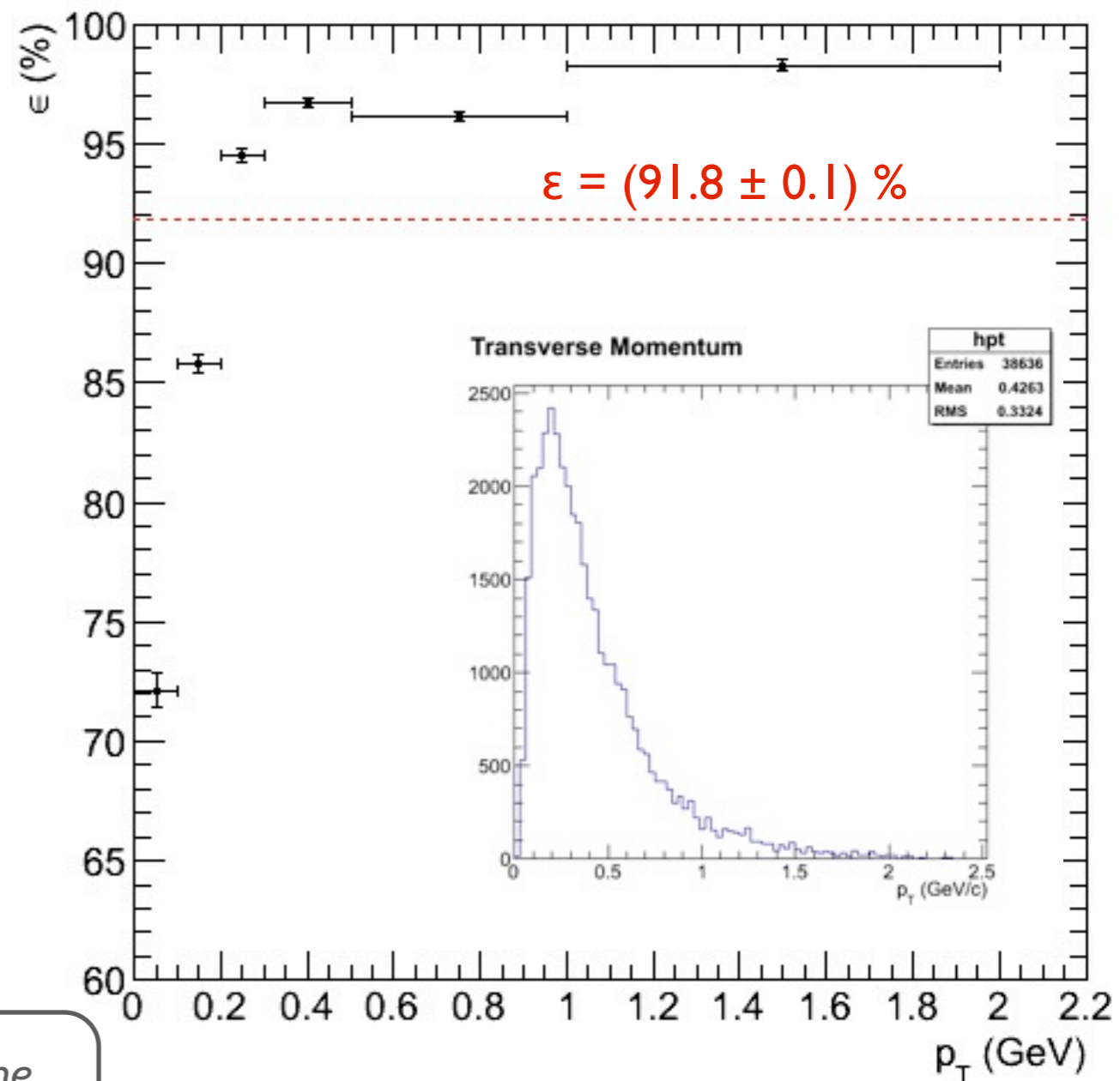
- ➔ *Efficiency* = $(91.8 \pm 0.1)\%$ = 38636/42072 PXDDigits

- strongly dependent on the *transverse momentum* (see next slide)

ROI Efficiency, p_T dependence

- **Efficiency** = $(91.8 \pm 0.1)\%$ = 38636/42072 PXDDigits
 - strongly dependent on the *transverse momentum*
- **Inefficiency** mostly due to failures in fitting the track and finding an intercept with the sensor planes
 - 94% of the times no intercept is found
 - increasing the size of ROI will not have a significant impact on the efficiency
 - 6% of the times a ROI is defined, 95% of which the sensor is the wrong one.

ROI efficiency - MCTrackFinder

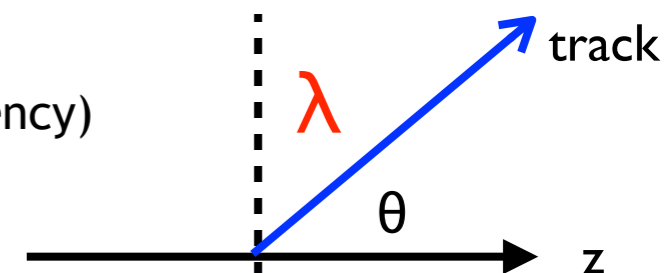


$$\epsilon = \frac{\# \text{ PXDDigits inside a ROI}}{\text{total \# PXDDigits of GFTrackCand}}$$

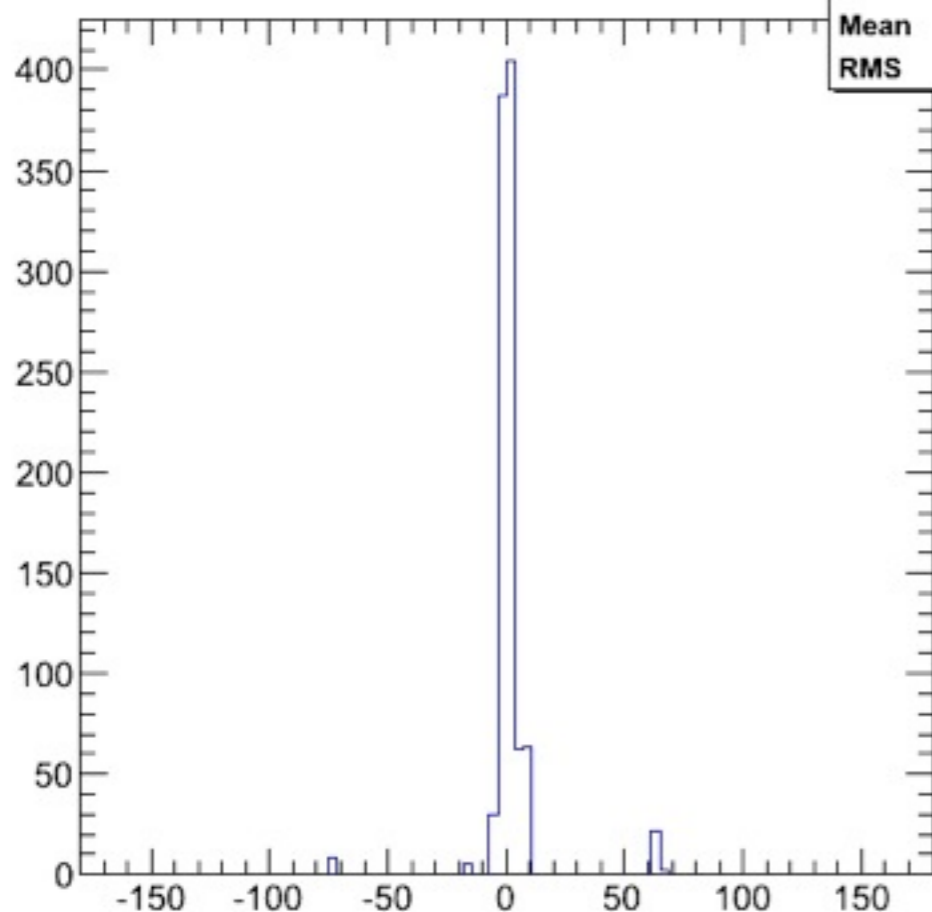
inefficiencies of the pattern recognition are factorized!!

What Hits are we Missing?

- ➔ mainly hits of low transverse-momentum tracks
- ➔ later hits of tracks looping on the plane $z \approx 0$ (~30% of the inefficiency)
- ➔ first hits of tracks at $\lambda \approx 0^\circ$ and $\lambda \approx 65^\circ$ (~70% of the inefficiency)

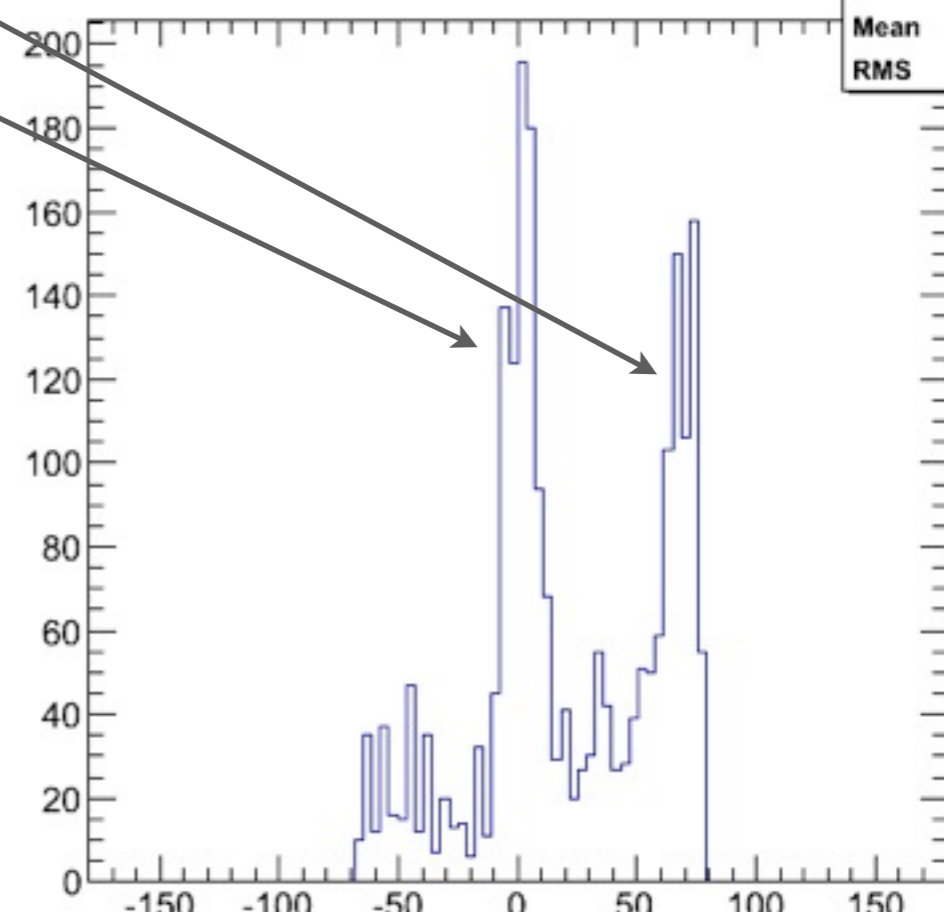


global time > 1 ns



λ

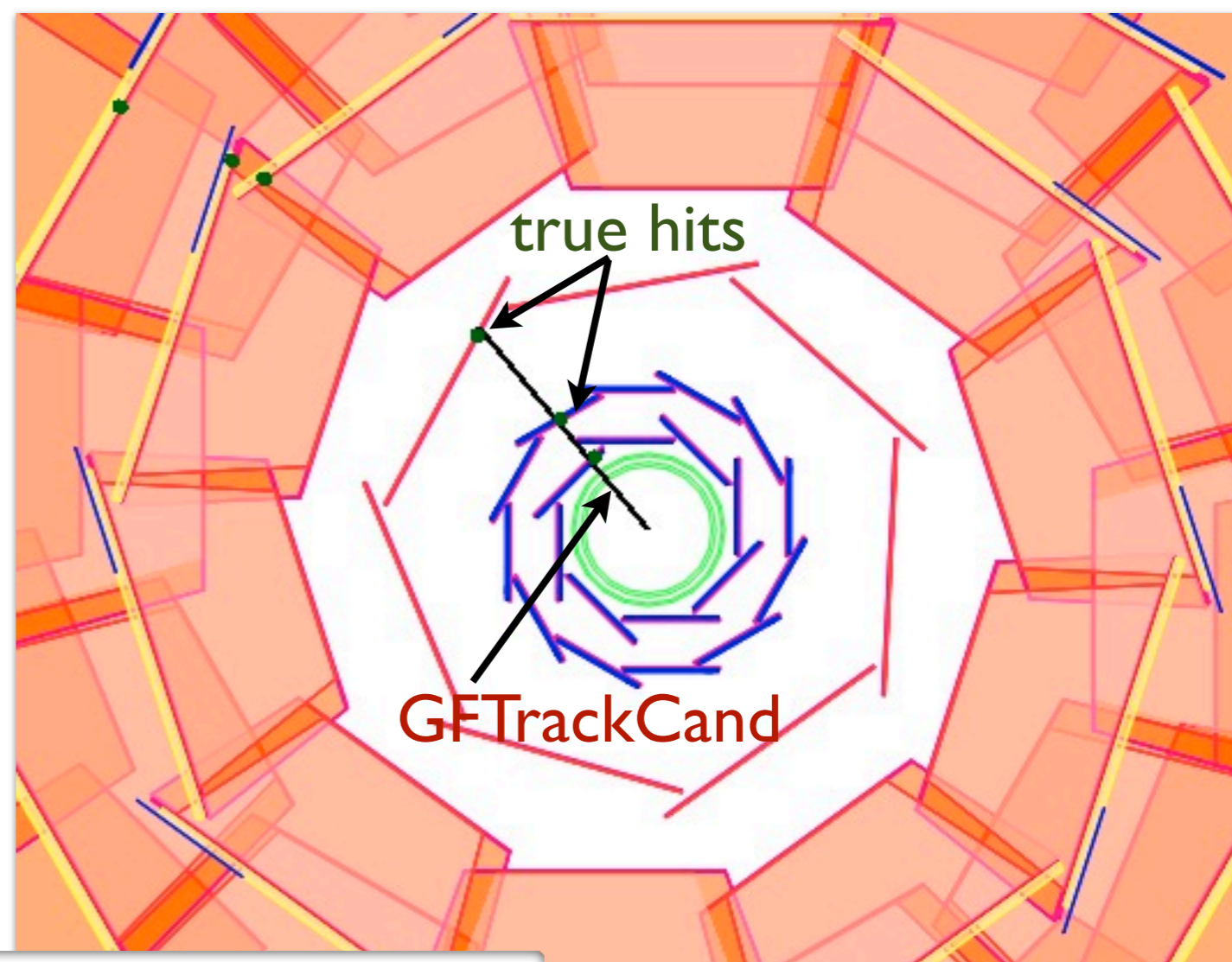
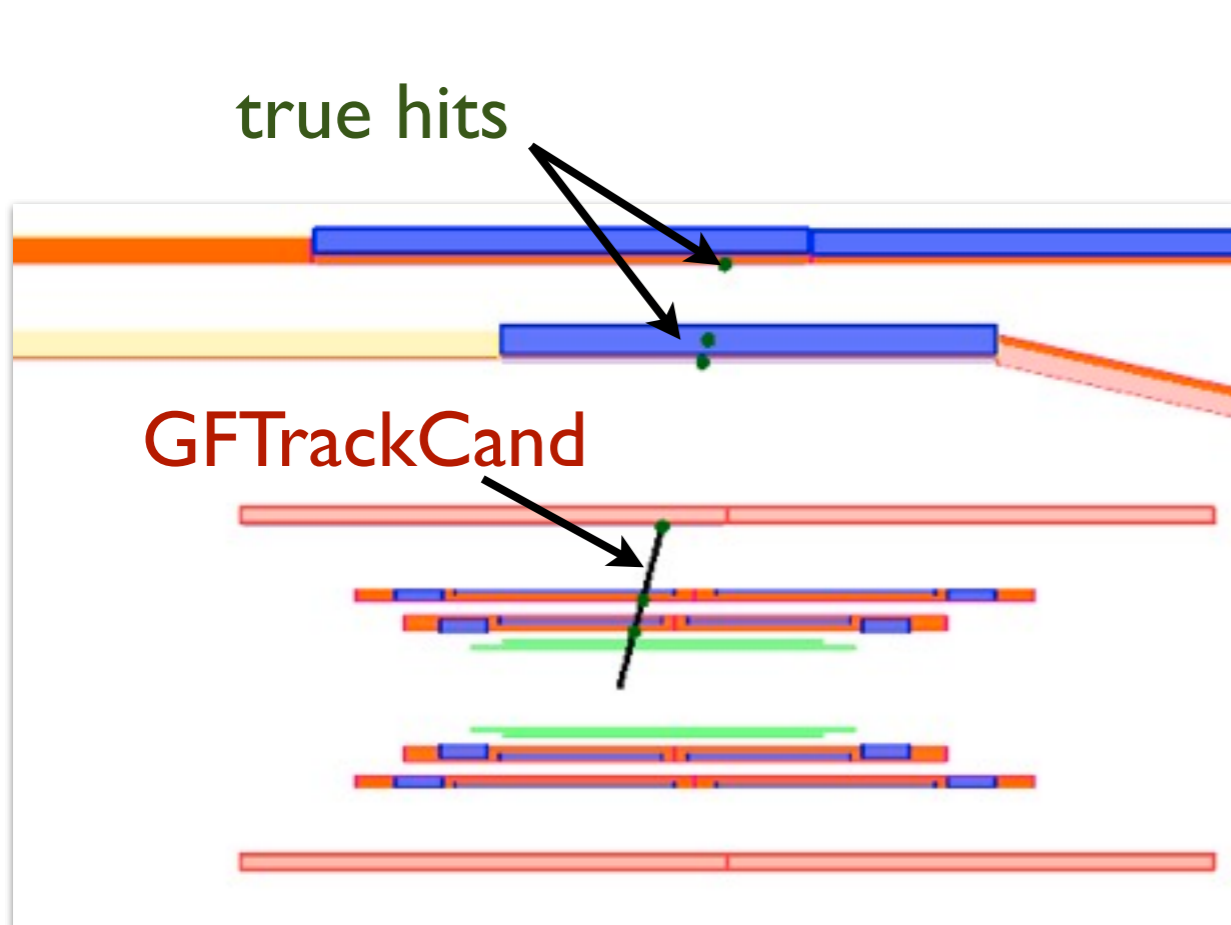
global time < 1 ns



λ

entries = tracks
for which the
fit has a "bad
track status" or
the intercept is
not found

Inefficiency due to *Bad Track Status*



```

GException thrown with excString:
RKTrackRep::RKutta ==> Do not get closer to plane!
in line: 1230 in file: /home/buildbot/externals/v00-04-01/src/genfit/RKTrackRep/RKTrackRep.cxx
with fatal flag 0

```

```

GException thrown with excString:
RKTrackRep::Extrap ==> maximum number of iterations exceeded
in line: 934 in file: /home/buildbot/externals/v00-04-01/src/genfit/RKTrackRep/RKTrackRep.cxx
with fatal flag 0

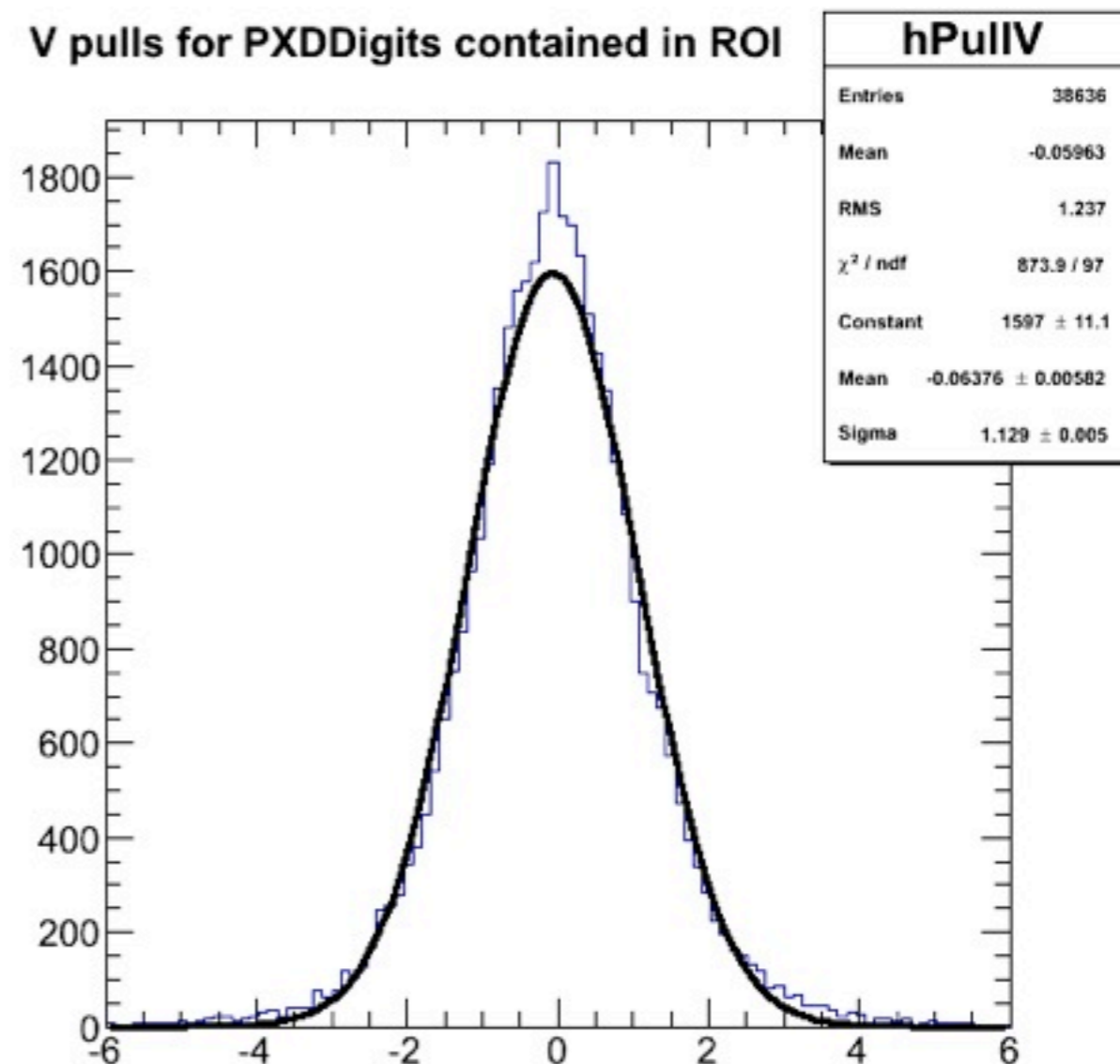
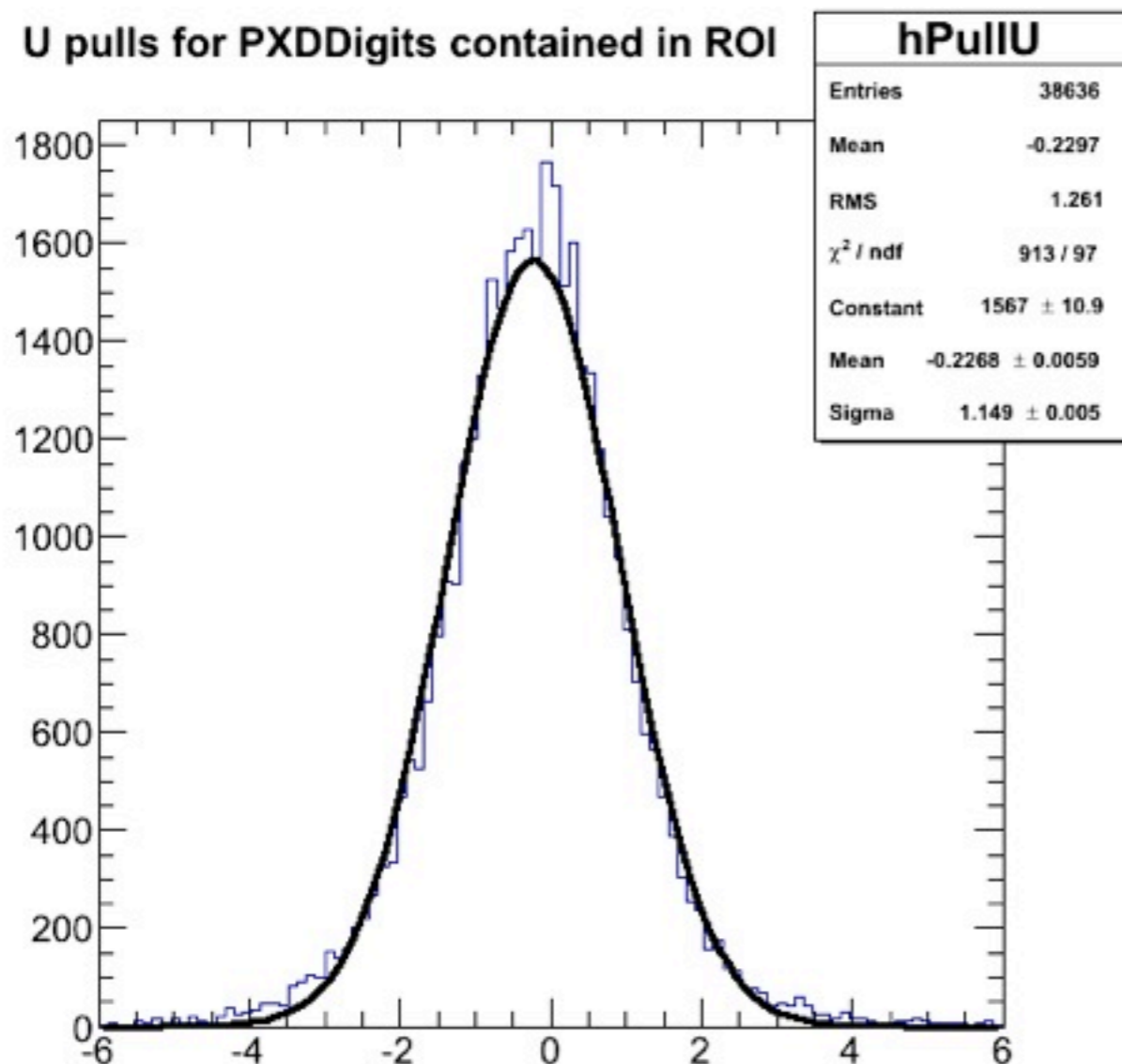
```

```

GException thrown with excString:
RKTrackRep::RKutta ==> momentum too low: 2.56996 MeV
in line: 1134 in file: /home/buildbot/externals/v00-04-01/src/genfit/RKTrackRep/RKTrackRep.cxx
with fatal flag 0
[WARNING] bad track status { module: PXDDataReduction }

```


ROI Definition, the U and V pulls



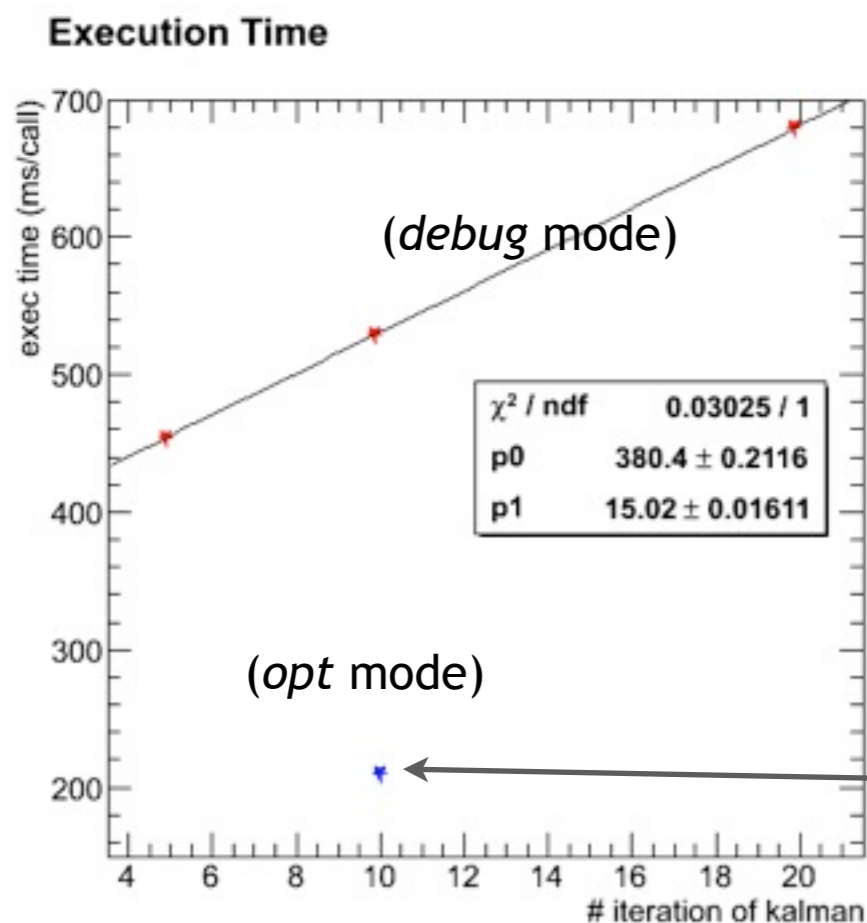
- ➔ pull = (intercept - center of the fired pixel) / (stat error on intercept)
- ➔ U (V) pull are slightly biased to negative values by 20% (5%) of the stat error
- ➔ U, V stat errors are underestimated by ~10-15%

Data Reduction Factor & Execution Time

➔ Data Reduction Factor = $\frac{\langle \# \text{ pixels in ROI/event} \rangle}{250 \cdot 768 \text{ pixels/module} \cdot 40 \text{ modules}}$

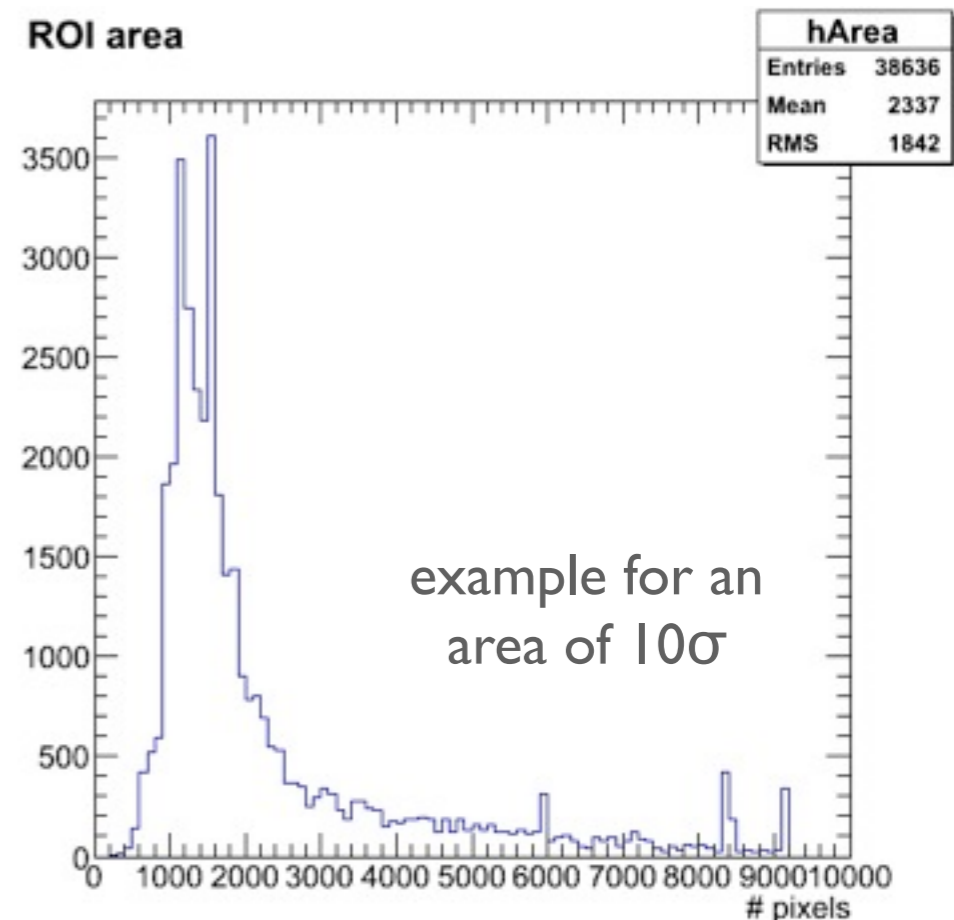
- ▶ 1.2% taking 10σ in each direction
- ▶ 0.4% taking 5σ in each direction

(to be tested with background)



➔ Execution time (code compiled in *debug* mode):

- ★ 45.5 ms/track with 5 iterations of the kalman filter
- ★ 53 ms/track with 10 iterations of the kalman filter
- ★ opt mode: 21 ms/track with 10 iterations
- ★ 68 ms/track with 20 iterations of the kalman filter



MCTrackFinder vs VXDTF

track candidates are built with the true SVT hits

track candidates produced by real pattern recognition

➔ We simulate 1k events using EvtGen and use the *VXDTF* as *pattern recognition*:

- ▶ ~8.5 tracks/event
- ▶ ~4.1 PXDDigits/track

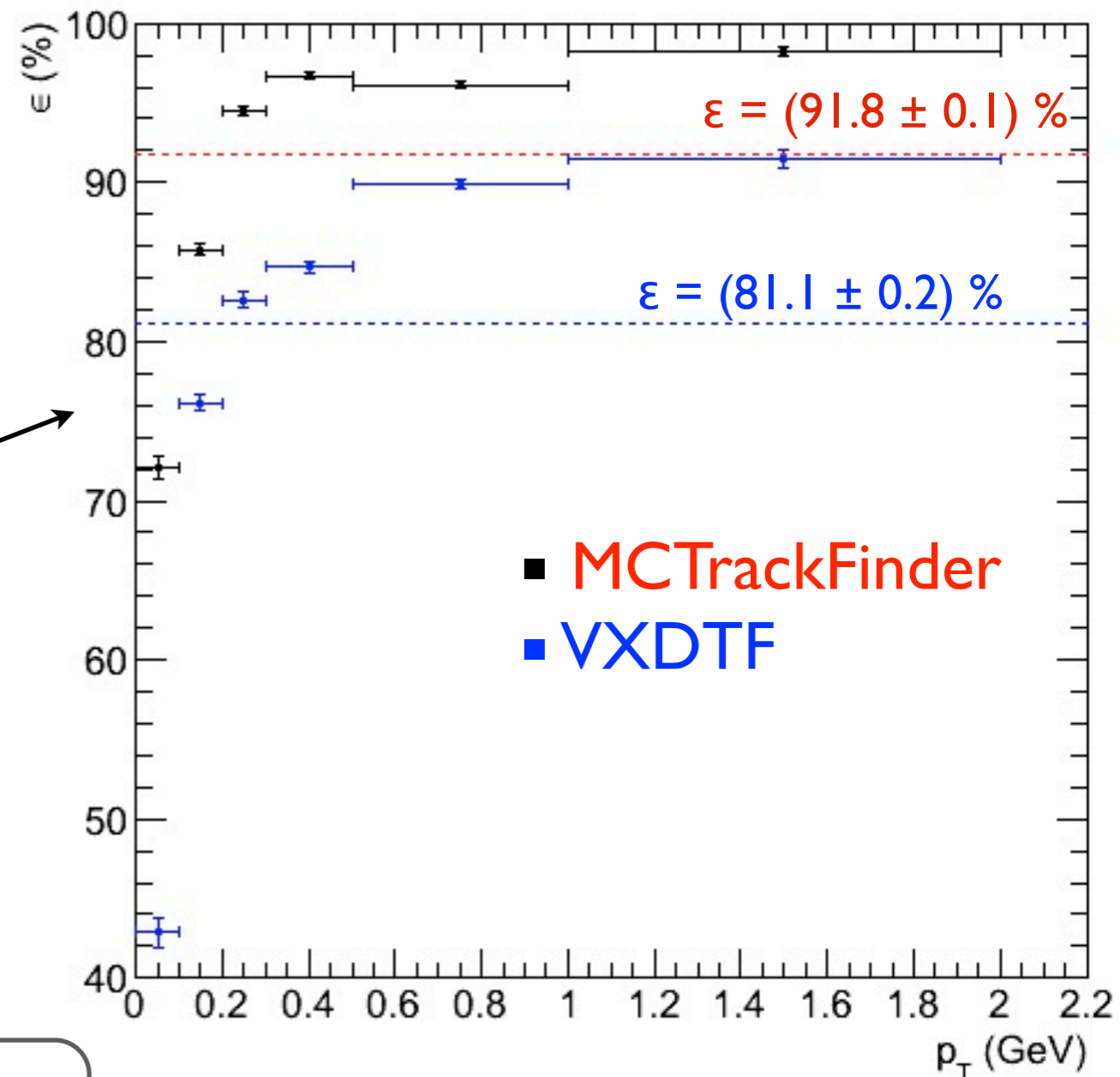
➔ **Efficiency = $(81.1 \pm 0.2)\%$** = 28388/34988 PXDDigits

- ▶ similar dependence on transverse momentum observed with MCTrackFinder

➔ Inefficiency mostly due to failures in fitting the track and finding an intercept with the sensor planes

- ▶ similar behavior observed with MCTrackFinder

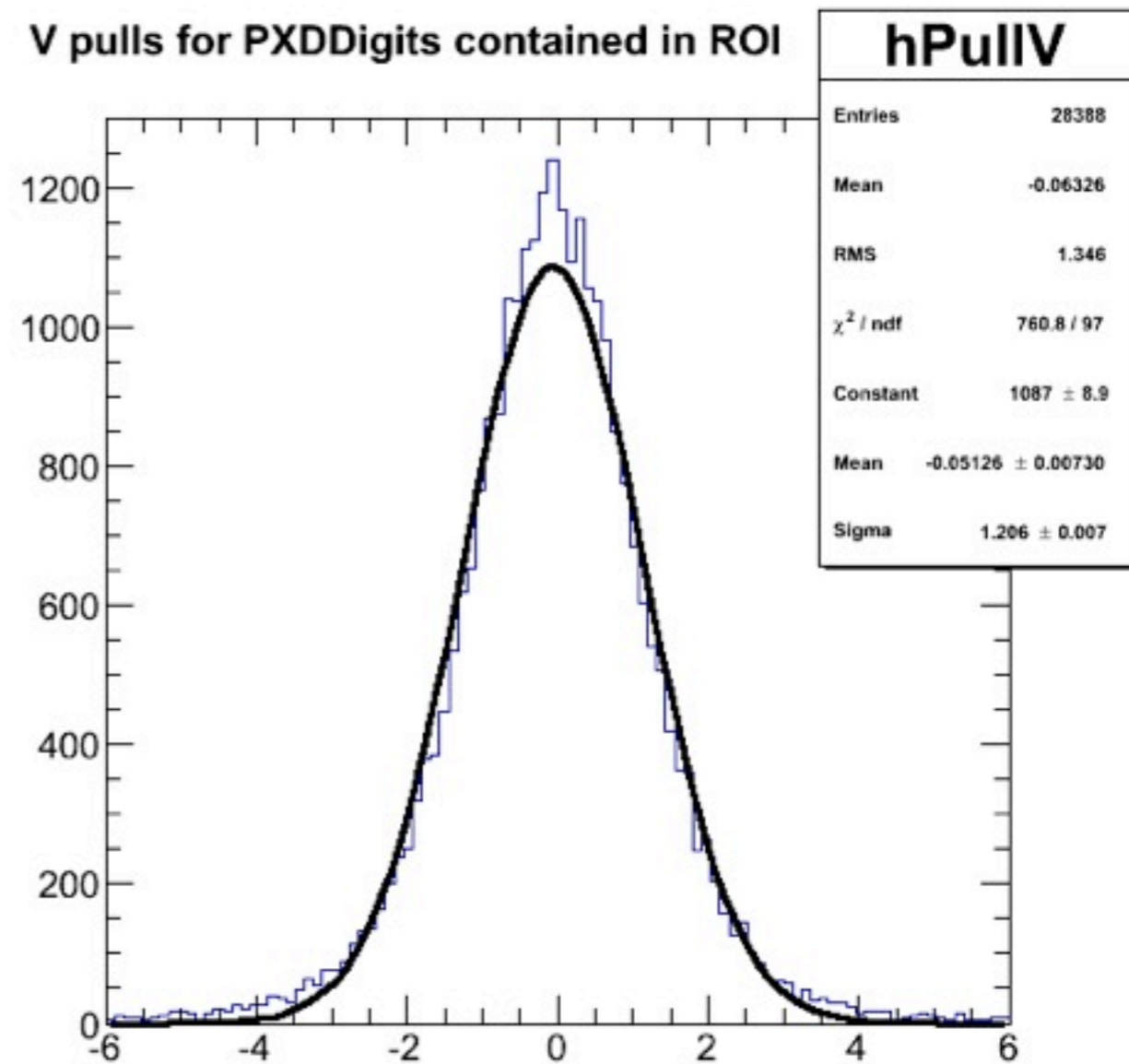
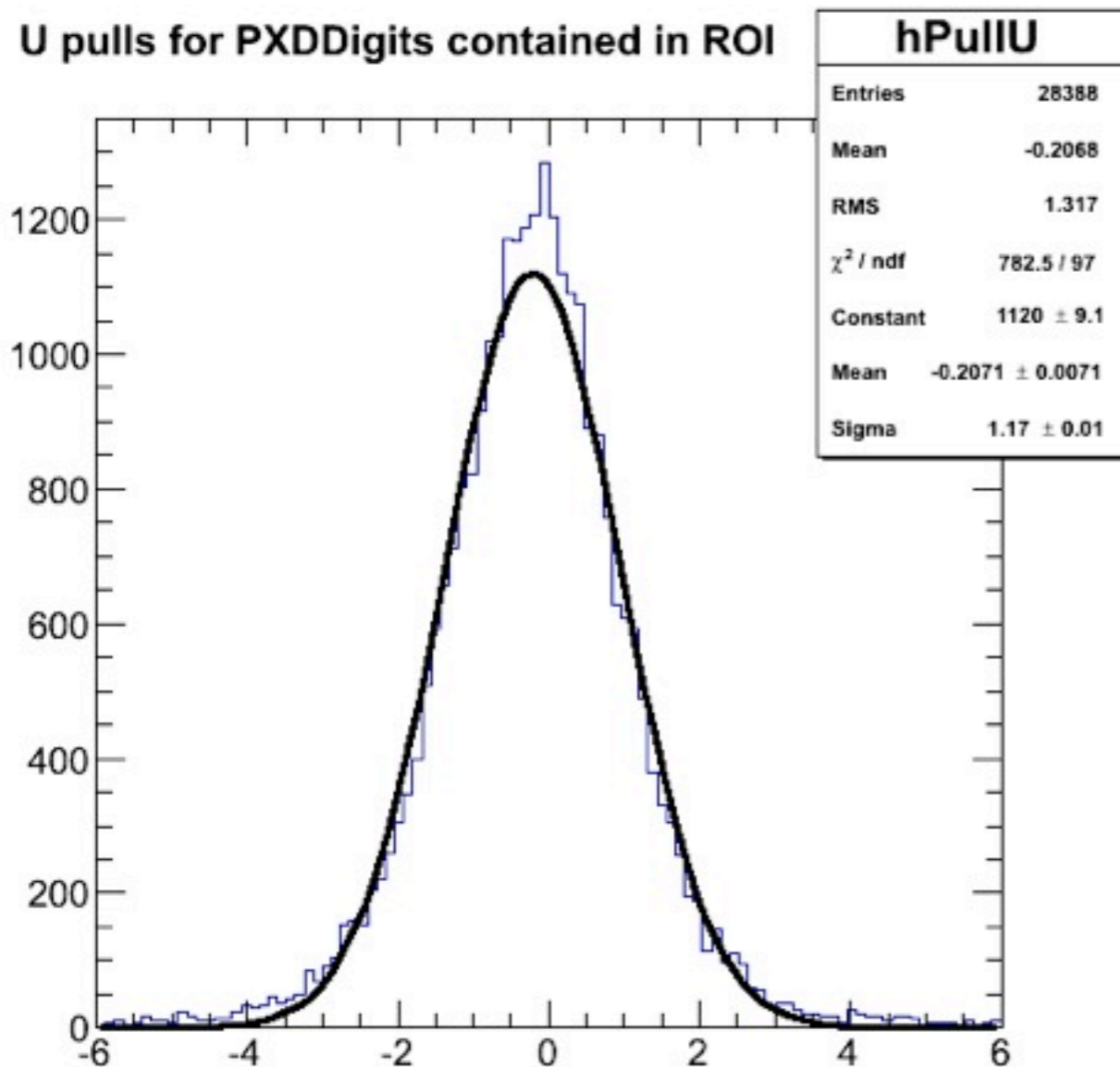
MCTrackFinder vs VXDTF



$$\epsilon = \frac{\# \text{ PXDDigits inside a ROI}}{\text{total \# PXDDigits of GFTrackCand}}$$

inefficiencies of the pattern recognition are factorized!!

ROIs with the VXDTF



- ➔ U (V) Pulls are negatively biased by 20% (5%) of the statistical error
- ➔ the statistical errors are underestimated by ~10-15%
- ➔ Data Reduction Factor = 0.8%
- ➔ Execution time = 35 ms/track (10 iterations of the kalman filter)

Performance for Beam Test

- ➔ We simulate 1k events using particleGun (2GeV e^- , no beam divergence) and use the *VXDTF* as *pattern recognition*:
 - ~0.78 tracks/event (0.83 with MCTrackFinder)
 - ~3.3 PXDDigits/track
- ➔ Efficiency = $(99.0 \pm 0.2)\%$ = 2588/2615 PXDDigits
- ➔ Data Reduction Factor = 0.05%
- ➔ Execution time = 2 ms/track

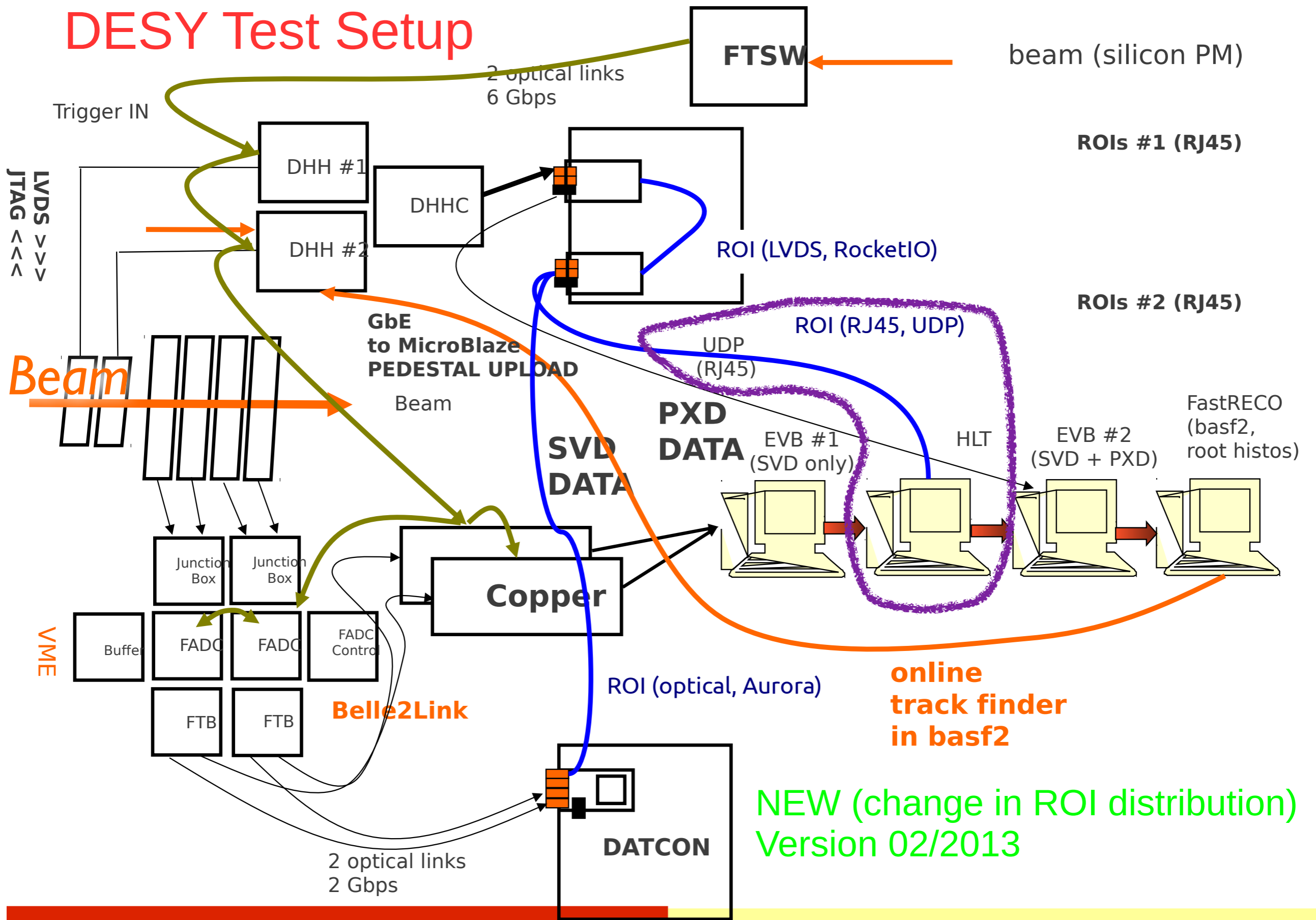
Conclusions

- ➔ PXDDataReduction Module is based on an pattern recognition module and a track-fit function whose performance influence the efficiency of the module
- ➔ The estimated ROI inefficiency of 20% (caveat emptor: pattern recognition inefficiency not included) is due failures in the fit and extrapolation of low transverse-momentum tracks
 - to increase the efficiency we should concentrate on the quality of the pattern recognition and of the fit of the track (genfit2 may help)
- ➔ The Data Reduction Factor is of the order of 1%, largely satisfactory
 - to be tested with background
- ➔ The performance of the module [test-beam geometry](#) are largely satisfactory from the points of view of efficiency, data reduction factor and execution time

Thank You!

backup-slides

DESY Test Setup



Parameters of the Study

Geometry:

```
geometry = register_module('Geometry')
param_geometry = {'Components': ['BeamPipe',
                                'Cryostat',
                                'HeavyMetalShield',
                                'MagneticField',
                                'PXD',
                                'SVD',
                                'SVD-Support',
                                'CDC',
                                ]}

geometry.param(param_geometry)
```

MC Track Finder:

```
mctrackfinder = register_module('MCTrackFinder')
mctrackfinder.logging.log_level = LogLevel.INFO
mctrackfinder.logging.debug_level = 101
param_mctrackfinder = {
    'UseCDCHits': 0,
    → 'UseSVDHits': 1,
    → 'UsePXDHits': 0,
    'MinimalNDF': 3,
    'UseClusters': 1,
    'WhichParticles': ['PXD', 'SVD'],
    'GFTrackCandidatesColName': 'mcTracks',
}
mctrackfinder.param(param_mctrackfinder)
```

VXD Track Finder:

```
vxdtf = register_module('VXDTF')
param_vxdtf = {
    'GFTrackCandidatesColName': 'vxdtfTracks',
    'tccMinState': [2],
    'tccMinLayer': [3],
    'reserveHitsThreshold': [0.6],
    → 'sectorSetup': ['evtNormSecHIGH_SVD', 'evtNormSecMED_SVD', 'evtNormSecLOW_SVD'],
    'calcQIType': 'circleFit',
    'tuneCircleFit': [0.001, 0.001, 0.00001],
    'filterOverlappingTCs': 'hopfield',
    'cleanOverlappingSet': True,
    'activateZigZagXY': [False, True, True],
    'TESTERexpandedTestingRoutines': True,
    'qiSmear': False,
}
vxdtf.param(param_vxdtf)
```

PXDDataReduction Module User Interface

maximum number of iteration of the kalman filter

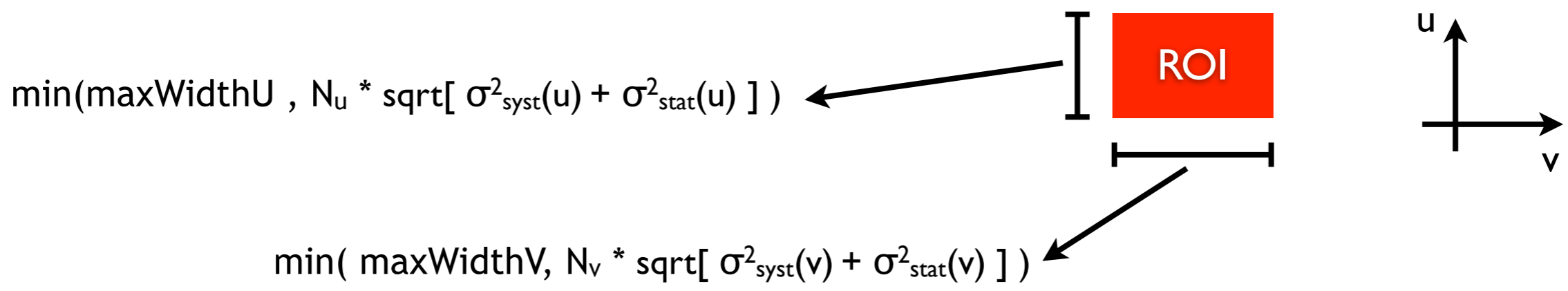
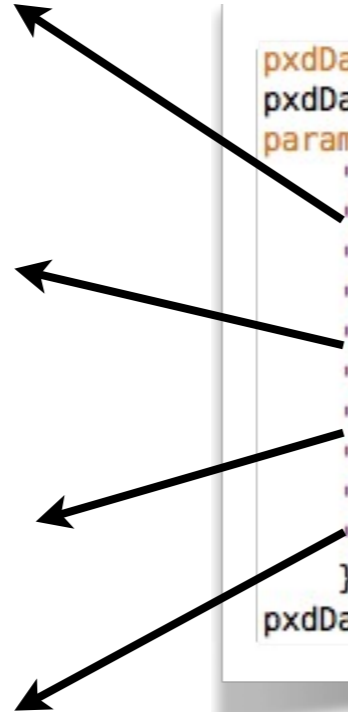
σ_{sys} to be added in quadrature to σ_{stat} from the extrapolation

$N_{u,v}$, the total number of $\sigma = \text{sqrt}(\sigma_{\text{syst}}^2 + \sigma_{\text{stat}}^2)$ to define the ROI

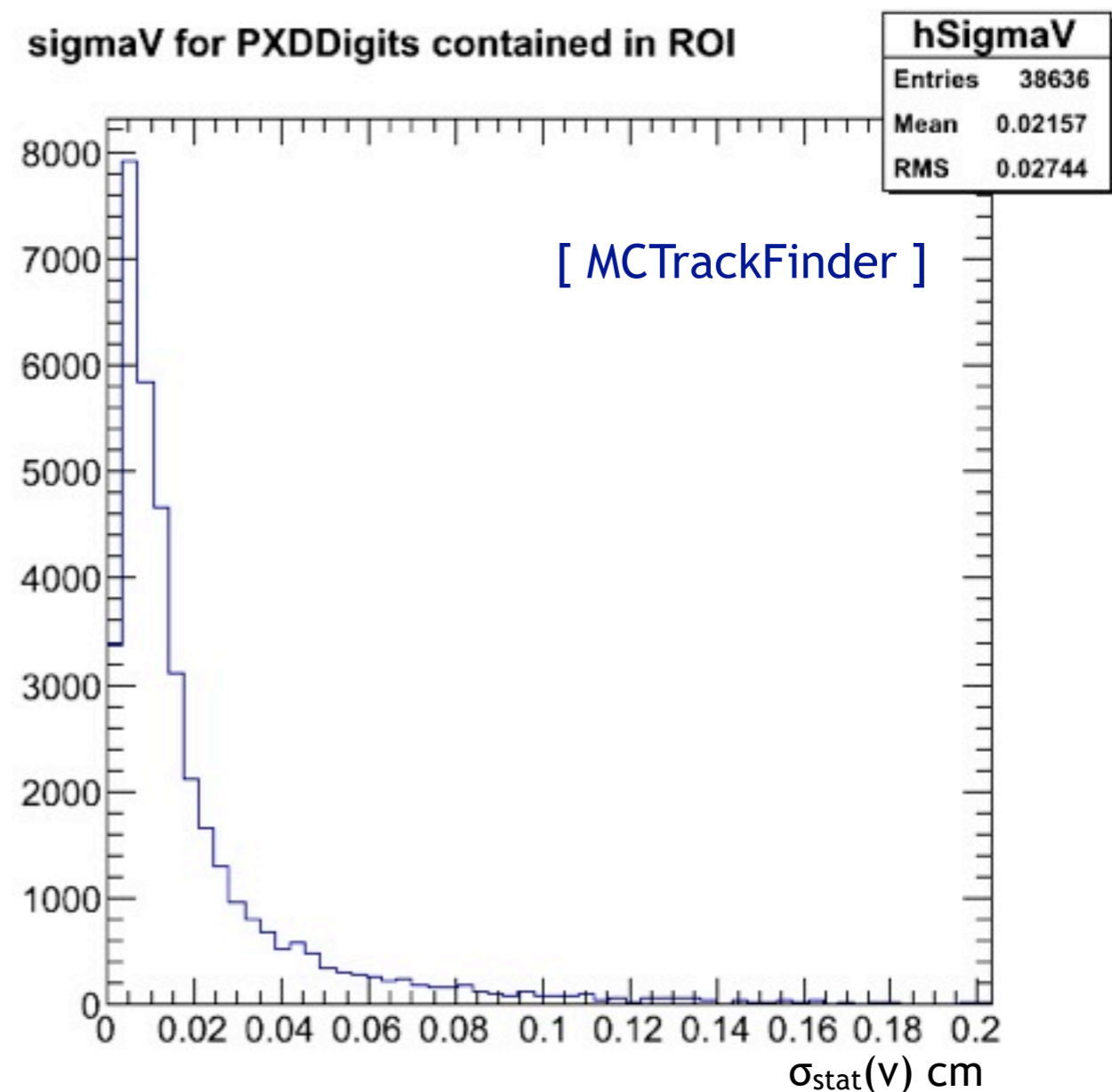
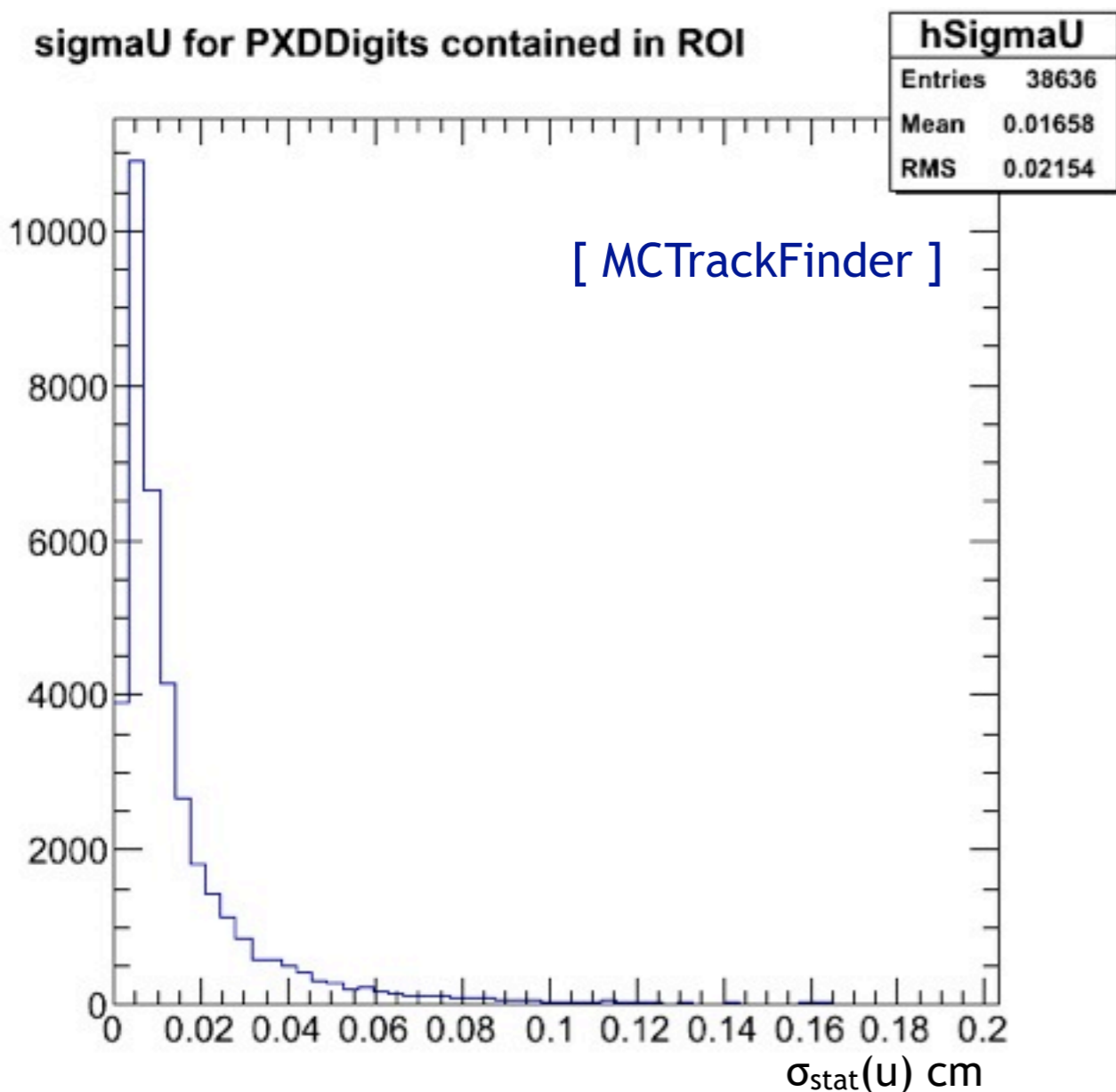
maximum allowed width of the ROI

```

pxdDataRed = register_module('PXDDataReduction')
pxdDataRed.logging.log_level = LogLevel.INFO
param_pxdDataRed = {
    'trackCandCollName': 'mcTracks',
    'numIterKalmanFilter': 10,
    'PXDIInterceptListName': 'PXDIIntercepts',
    'ROIListName': 'ROIs',
    'sigmaSystU': 0.02,
    'sigmaSystV': 0.02,
    'numSigmaTotU': 10,
    'numSigmaTotV': 10,
    'maxWidthU': 0.5,
    'maxWidthV': 0.5,
}
pxdDataRed.param(param_pxdDataRed)
    
```



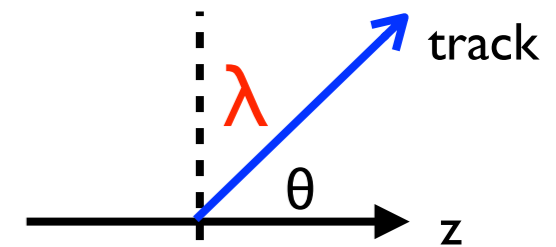
Statistical Error of the Extrapolation



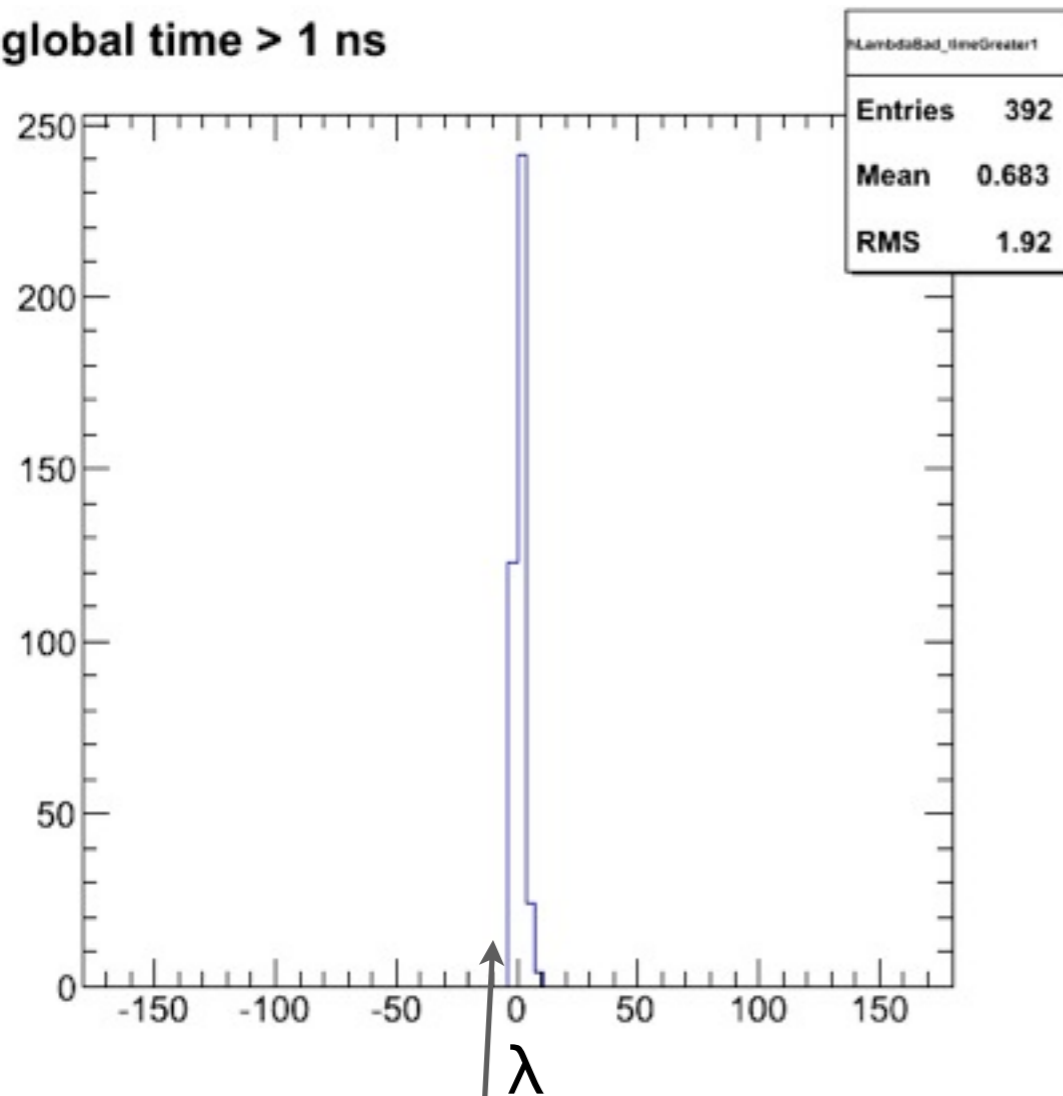
- ➔ with the VXDTF we observe similar statistical errors :
- in the U direction: mean = 0.015 cm (-10%), RMS = 0.021 cm (+13%)
 - in the V direction: mean = 0.020 cm (-7%), RMS = 0.031 cm (-5%)

Main Source of Inefficiency

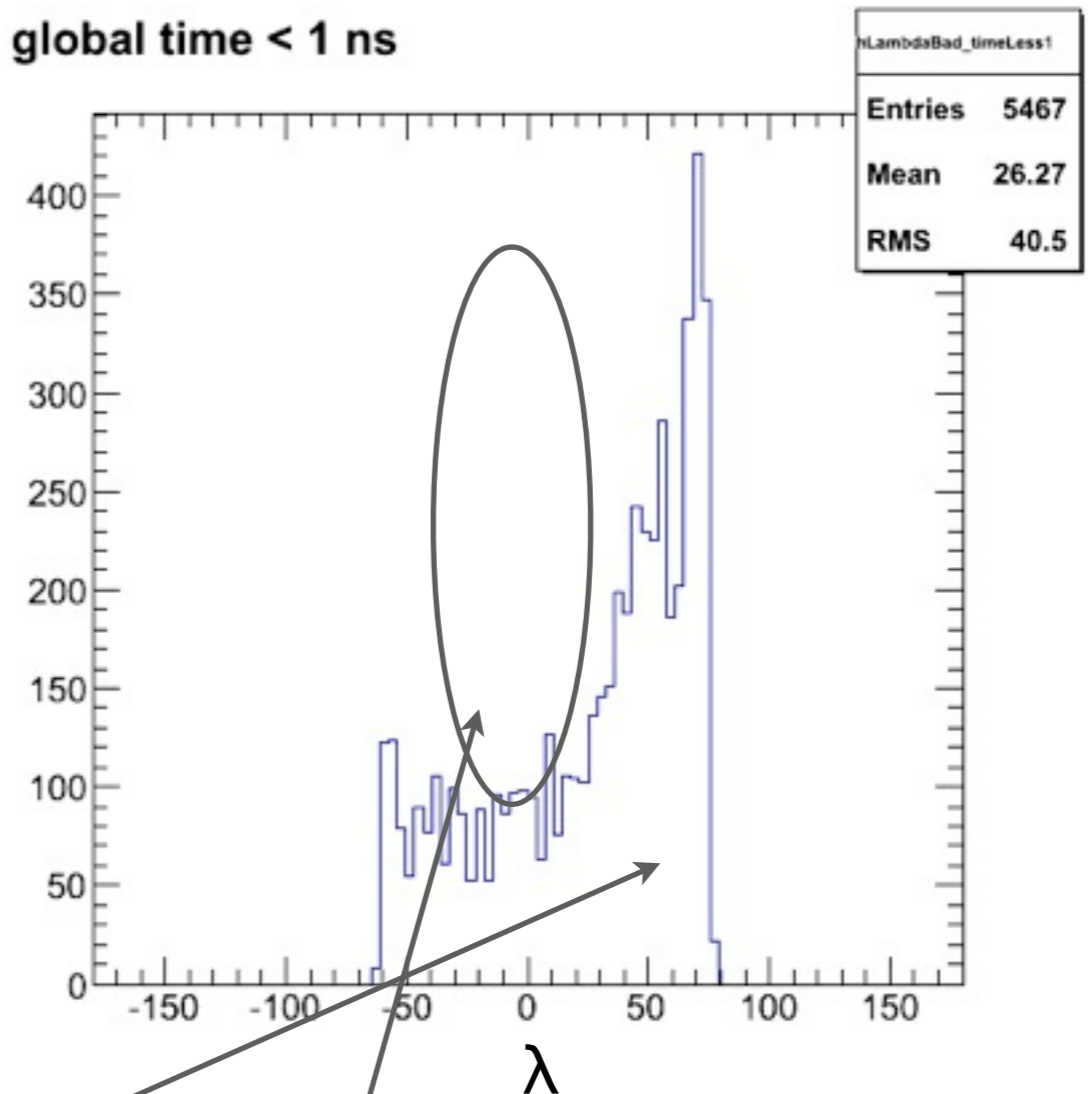
→ tracks for which the fit has a bad track status or the intercept is not found:



global time > 1 ns



global time < 1 ns



- mainly hits of low transverse-momentum tracks
- later hits of tracks looping on the plane $z \approx 0$ (7%)
- first hits of tracks $\lambda \approx 65^\circ$ (93%). First hits of track at $\lambda=0$ disappeared from the plot.